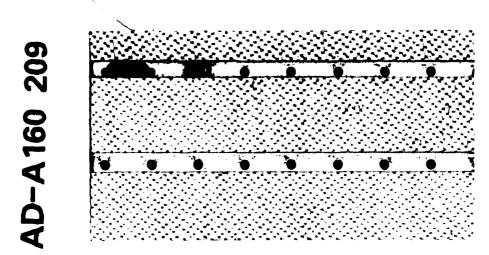


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LIMITING BEHAVIOR OF THE NORM OF PRODUCTS OF RANDOM MATRICES AND TWO PROBLEMS OF GEMAN-HWANG*

by

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LIMITING BEHAVIOR OF THE NORM OF PRODUCTS OF RANDOM MATRICES AND TWO PROBLEMS OF GEMAN-HWANG

bу

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ABSTRACT

In this paper, the authors proved that

$$\lim_{n\to\infty} ||(W\sqrt{n})^k|| \le (1+k)\sigma^k$$
, a.s.

where W: n x n is a square random matrix with i.i.d. entries and σ^2 is the variance of the entries of W. In proving the result, the authors assumed the existance of fourth monent of the entries of W.

Key words and Phrases: Spectral radius, limiting behavior, random matrices.



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1. INTRODUCTION

In the theory of large random matrices, how to dominate the norm of a random matrix is a very important problem. This is the reason why many authors are interested in this problem. For interesting works, see Geman (1980), Jonsson (1983), Silverstein (1984) and Yin-Bai-Krishnaiah (1984). In these papers, they consider the norm of a sample covariance matrix, with different moment requirements. The newest result of Yin-Bai-Krishnaiah requires only the existence of 4th moment.

In this paper, we consider a different type of random matrices, namely Wk, i.e. a power of a square random matrix with iid entries.

The first result in this paper (Theorem 2.1) is $\frac{1}{\lim_{n\to\infty} \left| \left(\frac{W}{\sqrt{n}} \right) \right|} \leq (1+k)\sigma^k, \text{ a.s. (n is the size of W),}$

here σ^2 is the variance of the entries of W. We assume only the existence of the 4-th moment of the entries of W. From this result it is easy to show that the spectral radius of W/ \sqrt{n} is not greater than σ with probability 1. This result is known only for iid N(0, σ^2) case.

In proving the above result, a new kind of graphs has to be discussed carefully, (\$3), and the truncation method used in Yin-Bai-Krishnaiah (1984) is also important here.

As applications of the above result, we have solved two open problems announced in the paper Geman-Hwang (1982). The solutions are in §5, §6 and §7.

LIMITING BEHAVIOR OF MATRIX PRODUCT NORM

In Sections 2-4, we will prove the following theorems.

Theorem 2.1. Let $\{w_{ij}: i=1,2,...,j=1,2,...\}$ be iid random variables, and W_n be the $n \times n$ matrix (w_{ij}) i,j = 1,2,...,n. Suppose

$$E w_{11} = 0$$
, $E w_{11}^2 = \sigma^2$, $E w_{11}^4 < \infty$. (2.1)

Then, for any positive integer k, we have

$$\lim_{n\to\infty}\sup \left|\left|\left(\frac{W_n}{\sqrt{n}}\right)^k\right|\right| \leq (k+1)\sigma^k \quad a.s. \tag{2.2}$$

Here | | A | | denotes the operator norm of the matrix A.

Denote by $\lambda_i(A)$, $i=1,2,\ldots,n$, the n eigenvalues of the $n\times n$ matrix A. We have

Theorem 2.2. Under the same conditions as in Theorem 2.1, we have

$$\lim_{n\to\infty}\sup_{1\leq i\leq n}|\lambda_i(\frac{w_n}{\sqrt{n}})|\leq\sigma\quad a.s.$$

Theorem 2.2 can be easily deduced from Theorem 2.1 as follows: For any integer $k \ge 1$, by Theorem 2.1,

$$\lim_{n\to\infty} \sup_{1\leq i\leq n} \max_{1\leq i\leq n} \left| \lambda_i \left(\frac{w_n}{\sqrt{n}} \right) \right| = \lim_{n\to\infty} \sup_{1\leq i\leq n} \max_{1\leq i\leq n} \left| \lambda_i \left[\left(\frac{w_n}{\sqrt{n}} \right) \right] \right|^{1/k}$$

$$\leq \lim_{n\to\infty} \sup_{n\to\infty} \left| \left| \left(\frac{w_n}{\sqrt{n}} \right)^k \right| \right|^{1/k} \leq (k+1)^{1/k} \sigma. \quad a.s.$$

Letting $k \rightarrow \infty$ we get Theorem 2.2.

3. SOME LEMMAS

At first we state the following lemma which can be found in Yin-Bai-Krishnaiah (1984).

Truncation Lemma. Let r be a number in the interval $[\frac{1}{2}, 2]$, $\{w_{ij}: i,j=1,2,\ldots\}$ be a set of iid random variables with E $w_{11}=0$, $\mathbb{E}|w_{11}|^{2/r}<\infty$. For each n, let W_n denote the p×n matrix whose (i,j)-entry is w_{ij} , here p = p(n) satisfies p/n \longrightarrow y \in $(0,\infty)$, as $n \to \infty$.

Then there exists a sequence of positive numbers $\delta = \delta_n$ such that

- 1. $\delta \rightarrow 0$, as $n \rightarrow \infty$,
- 2. $P(W_n \neq \hat{W}_n, i.o.) = 0$; here \hat{W}_n is the p×n matrix, with the (i,j) entry $\hat{W}_{ijn} = W_{ij}^{-1} \{ |w_{ij}| < \delta n^r \}$,

and the convergence speed of δ to zero can be slower than any preassigned speed.

In order to prove Theorem 2.1, we need some combinatorics. Let $i_1, i_2, \ldots, i_{2km}$ be a sequence, we define a multigraph $\Gamma(k, m; i_1, \ldots, i_{2km})$ as follows:

- 1. The vertices of this graph are $i_1, i_2, \ldots, i_{2km}$. Some of them may be equal.
- 2. There are 2km edges e_1, e_2, \dots, e_{2km} . The ends of e_a are i_a and i_{a+1} $(i_{2km+1} = i_1)$. Any two of these edges are different even when they have the same end sets. Sometimes we write $i_a i_{a+1}$ instead of e_a .
- 3. To each edge e_a there corresponds a number $dir(e_a)$, called the <u>direction</u> of e_a , such that

dir
$$(e_a) = \begin{cases} +1, & \text{if } [(a-1)/k] \text{ is even,} \\ -1, & \text{if } [(a-1)/k] \text{ is odd.} \end{cases}$$

Two edges $e_a = i_a i_{a+1}$, $e_b = i_b i_{b+1}$ are said to be coincident, if either $i_a = i_b$, $i_{a+1} = i_{b+1}$ and $dir(e_a) = dir(e_b)$, or $i_a = i_{b+1}$, $i_{a+1} = i_b$ and $dir(e_a) = -dir(e_b)$. A chainisa subgraph with vertex set $\{i_a, i_{a+1}, \dots, i_b\}$ $(1 \le a < b \le 2mk+1)$ and edge set $\{e_a, e_{a+1}, \dots, e_{b-1}\}$. We will denote such a chain by $i_a i_{a+1} \cdots i_b$.

In the graph $\Gamma(k,m;\ i_1,i_2,\ldots,i_{2km})$, we classify the edges as follows.

- 1. An edge $i_{a-1}i_a$ is called an innovation if i_a is new, i.e. $i_a \neq i_1, \dots, i_a \neq i_a$. The set of all innovations will be denoted by I.
- 2. Let S be the set of all edges $i_{a-1}i_a$ which coincides with an innovation, and for any b < a, $i_{h-1}i_b$ does not coincide with that innovation.
 - 3. All other edges consist a set called T.

If $i_{a}i_{a+1}$, $i_{b}i_{b+1}$ are two edges satisfying the following properties:

- (1) b < a;
- (2) $i_{b}i_{b+1}$ is single up to i_{a} , i.e. it does not coincide with any edge of the chain $i_{1}i_{2}...i_{a}$.
- (3) Either $i_b = i_a$ and $dir(i_b i_{b+1}) = dir(i_a i_{a+1})$, or $i_{b+1} = i_a$ and $dir(i_b i_{b+1}) = -dir(i_a, i_{a+1})$, then we say that $i_a i_{a+1}$ is coincidable with $i_b i_{b+1}$.

An edge of S is called singular if it is coincidable with just one innovation.

An edge of S is called regular if it is not singular, i.e. it is coincidable with more than one edge.

The proofs of Lemma 3.1, 3.2, 3.3 below are similar to the proofs of Lemma 3.1, 3.2, 3.3 in Yin-Bai-Krishnaiah (1984).

Lemma 3.1. If in the chain $i_a i_{a+1} \cdots i_b$, $i_a i_{a+1}$ is single up to i_b and i_b has been visited by $i_1 i_2 \cdots i_a$ then $i_a i_{a+1} \cdots i_b$ contains an edge of T.

Lemma 3.2. Let t be the number of equivalence classes of T under the equivalence relation "coincidence". Then if $i_a i_{a+1}$ is a regular edge of S, the number of edges with which $i_a i_{a+1}$ is coincidable is not greater than t+1.

Lemma 3.3. The number of regular edges of S is not greater than twice the number of edges in T.

The chains $L_1 = i_1 i_2 \cdots i_k i_{k+1}$, $L_2 = i_{k+1} i_{k+2} \cdots i_{2k+1}, \cdots$, $L_{2m} = i_{(2m-1)k+1} i_{(2m-1)k+2} \cdots i_{2m} i_1$ are called segments.

Lemma 3.4. Let ℓ be the number of innovations. Then the number of different ways to appoint the 2km edges to be of I, or S, or T, does not exceed $\binom{2km}{2\ell}$ $\binom{k+1}{2\ell}$.

Proof. Since the number of innovations are ℓ , the numbers of S and T must be ℓ and $2km-2\ell$, respectively. So there are $\binom{2km}{2\ell}$ different ways to select $2km-2\ell$ edges from the 2km edges which are appointed to be of T, and the others to be of I or of S.

Now consider a segment L_c . Note that every edge in the same segment has the same direction. Suppose that L_c contains μ_c edges of T. Then L_c is split by these μ_c T-edges into at most μ_c + 1 subchains consisting of consecutive edges of I \bigcup S. Let the lengths of these subchains be $v_1, v_2, \ldots, v_{\mu_c+1}$, respectively (if there are less than μ_c + 1 such chains, then some v_i at the rear part of this list are zero). Consider the i-th subchain with v_i edges. It is evident that if some edge in this chain is of I, then the next one (if any) must be of I because of the same direction of them. So there are only v_1 + 1 possible appointments for this chain, namely, III ... I, SII ... I, SSI ... I, SSS ... S. So for the whole segment L_c , there

are at most $\prod_{i=1}^{\mu_C+1} (v_i+1) \le (k+1)^{\mu_C+1}$ ways to appoint the $k-\mu_C$ non-T the 21 non-T edges to be of I or of S.

4. PROOF OF THEOREM 2.1

Now we apply the truncation lemma for $r=\frac{1}{2}$ and p(n)=n. We need only to prove

$$\limsup_{n\to\infty} \left| \left| \left(\frac{\hat{W}_n}{\sqrt{n}} \right)^k \right| \right| \leq (k+1)\sigma^k. \quad a.s. \tag{4.1}$$

Define $\tilde{w}_{ijn} = \hat{w}_{ijn} - E \hat{w}_{ijn}$ and define $\tilde{w}_{n} = (\tilde{w}_{ijn})$, i,j = 1,2,...,n. We shall prove that for any k > 1

$$\lim_{n\to\infty}\sup \left|\left|\left(\frac{\widetilde{w}_n}{\sqrt{n}}\right)^k\right|\right| \leq (k+1)\sigma^k \quad a.s. \tag{4.2}$$

If (4.2) holds for any $k \ge 1$, since

$$\left| \left| \left| \left(\frac{\widehat{w}_{n}}{\sqrt{n}} \right)^{k} \right| \right| - \left| \left| \left(\frac{\widetilde{w}_{n}}{\sqrt{n}} \right)^{k} \right| \right| \leq \left| \left| \left(\frac{\widehat{w}_{n}}{\sqrt{n}} \right)^{k} - \left(\frac{\widetilde{w}_{n}}{\sqrt{n}} \right)^{k} \right| \right|$$

$$\leq \sum_{k=0}^{k-1} \left| \left| \left(\frac{\widehat{w}_{n}}{\sqrt{n}} \right)^{k} \right| \left| \left| \frac{\widehat{w}_{n}}{\sqrt{n}} - \frac{\widetilde{w}_{n}}{\sqrt{n}} \right| \right| \left| \left| \left(\frac{\widetilde{w}_{n}}{\sqrt{n}} \right)^{k-\ell-1} \right| \right|$$

and

$$\left|\left|\frac{\hat{W}_{n}}{\sqrt{n}} - \frac{\tilde{W}_{n}}{\sqrt{n}}\right|\right| = \frac{\left|E \ W_{11n}\right|}{\sqrt{n}} \quad \left|\left|\left[\begin{array}{c} 1, 1, \dots, 1 \\ 1, 1, \dots, 1 \\ 1, 1, \dots, 1 \end{array}\right]\right|\right| = \left|E \ W_{11n}\right| \longrightarrow 0,$$

by (4.2) we obtain

$$\lim_{n\to\infty} \sup_{l\to\infty} \left| \left| \left| \left(\frac{\widehat{w}_n}{\sqrt{n}} \right)^k \right| \right| - \left| \left(\frac{\widetilde{w}_n}{\sqrt{n}} \right)^k \right| \right|$$

$$\leq \lim_{n\to\infty} \sup_{\ell=0}^{k-1} \left| \left| \left(\frac{\widehat{w}_n}{\sqrt{n}} \right)^\ell \right| \left| E w_{11n} \right| (k-\ell)^{\sigma^{k-\ell-1}}$$
(4.3)

from which and by induction we can deduce (4.1). Hence to prove Theorem 2.1, we need only to prove (4.2).

For saving notations, we can assume that \textbf{W}_n is an $n\times n$ matrix with iid random entries $\textbf{w}_{i\,i}$, such that

$$E w_{11} = 0, |w_{11}| \le \delta \sqrt{n}, E w_{11}^2 \le 1.$$
 (4.4)

Here, without any loss, we suppose σ = 1, and instead of 2δ we write δ .

Under the condition (4.4), it is easy to see that

$$E|w_{11}^{\ell}| \leq \begin{cases} (\delta\sqrt{n})^{\ell-2}, & \text{for } \ell \geq 2, \\ d(\delta\sqrt{n})^{\ell-3}, & \text{for } \ell \geq 3. \end{cases}$$

$$(4.5)$$

It is enough to show that for any number z > (1+k)

$$\sum_{n=1}^{\infty} P(\left|\left(\frac{W_n}{\sqrt{n}}\right)^k\right| \right| \geq z) < \infty.$$
 (4.6)

But since

$$\left|\left|\left(\frac{W_{n}}{\sqrt{n}}\right)^{k}\right|\right|^{2m} \leq \left(\lambda_{\max}\left(\left[\left(\frac{W_{n}}{\sqrt{n}}\right)^{k}\right]^{T}\left(\frac{W_{n}}{\sqrt{n}}\right)^{k}\right)\right)^{m}$$

$$\leq \operatorname{tr}\left\{\left(\frac{W_{n}}{\sqrt{n}}\right)^{k}\left[\left(\frac{W_{n}}{\sqrt{n}}\right)^{k}\right]^{T}\right\}^{m}$$

For any sequence m = m(n) of positive integers,

$$\sum_{n=1}^{\infty} P(||(w_n / \sqrt{n})^k|| \ge z)$$

$$\leq \sum_{n=1}^{\infty} P(tr(w_n^k(w_n^k)^T)^m \ge z^{2m} n^{mk})$$

$$\leq \sum_{n=1}^{\infty} z^{-2m} n^{-mk} E tr(w_n^k(w_n^k)^T)^m.$$

And we need only to show that for some positive integers m = m(n),

$$\sum_{n=1}^{\infty} z^{-2m} n^{-mk} E tr(W_n^k(W_n^k)^T)^m < \infty.$$
 (4.7)

We have

$$E_n = E \operatorname{tr}(W_n^k(W_n^k)^T)^m = \sum E(w_{i_1 i_2} w_{i_2 i_3} \dots w_{i_k i_{k+1}})(w_{i_{k+2} i_{k+1}} w_{i_{k+3} i_{k+2}} \dots$$

$$\cdots w_{i_{2k+1}i_{2k}}) \cdots (w_{i_{(2m-1)k+2}i_{(2m-1)k+1}} \cdots w_{i_{2mk+1}i_{2mk}}).$$

Here, $i_1, i_2, \ldots, i_{2mk}$ run over $\{1, 2, \ldots, n\}$ and $i_{2mk+1} = i_1$. For each $i_1, i_2, \ldots, i_{2mk}$ we can define a graph $\Gamma(k, m)$ as in Section 3.

Now we estimate the expectation

$$m-1$$
 k
 $M = E$ Π Π $(w_i w_{i2ak+b+1} w_{i(2a+1)k+b+1} w_{i(2a+1)k+b})$.

where $(i_1, i_2, \dots, i_{2km})$ forms a canonical graph (i.e. $i_a \le \max(i_1, \dots, i_{a-1}) \forall a$, and $i_1 = 1$). Suppose that in the graph $\Gamma(k, m; i_1, i_2, \dots, i_{2km})$

- 1. There are & innovations
- 2. There are μ innovations which coincide with some T-edges, and the number of T-edges which coincide with the i-th innovation of this kind is n_i , $i=1,2,\ldots,\mu$.
- 3. There are t equivalence classes of T-edges split by the relation "being coincident".
- 4. These classes in 3 which do not contain any innovation have $m_1, m_2, \ldots, m_{t-\mu}$ edges respectively. $(m_i \ge 2, i = 1, 2, \ldots, t-\mu)$. It is easy to see that

$$M = (E w_{11}^{2})^{\ell-\mu} \prod_{j=1}^{\mu} E w_{11}^{n_{j}+2} \prod_{j=1}^{t-\mu} E w_{11}^{m_{j}},$$

and

$$2(\ell-\mu) + \sum_{i=1}^{\mu} (n_i+2) + \sum_{j=1}^{t-\mu} m_j = 2mk.$$

So by (4.4) and (4.5), we obtain

$$|\mathbf{M}| < \mathbf{d}^{\mu} \left(\delta \sqrt{\mathbf{n}}\right)^{2km-2\ell-t} \le \mathbf{m}^{t} \left(\delta \sqrt{\mathbf{n}}\right)^{2km-2\ell-t} \tag{4.8}$$

for n, hence for m, large enough.

Now we estimate the sum E of all expectations whose graphs $\Gamma(k,m)$ do not have single edges.

Let ℓ denote the number of innovations of the graph $\Gamma(k,m)$. Then there are ℓ S-edges and $2km-2\ell$ T-edges. For a fixed canonical graph $\Gamma(k,m)$ with ℓ innovations, there are n! / $(n-\ell-1)! \le n^{\ell+1}$ different graphs which correspond to this canonical graph.

By Lemma 3.4, there are at most $\binom{2 \text{km}}{20} (k+1)^{2 \text{km}-2l+2m}$ different ways to appoint the 2km edges to be of I or of S or of T.

Let t denote the number of noncoincident T-edges. Because our graphs do not have single throughout edges, we have $\ell \leq mk$ $1 \le t \le 2km-2\ell$ if $\ell \le mk-1$.

Next we bound the number of different ways to appoint each edge in a canonical graph with given positions of the & innovations, & S-edges and 2km-2l T-edges and with t different T-edges. Since each edge is an element of the left-upper $2km \times 2km$ submatrix of $W_{\underline{a}}$ so there are at most ((2km)2) t2km-2k different ways to appoint the t different T-edges into their 2km-2l different positions.

Each innovation in a canonical graph is uniquely determined by the edges before it, and so is each singular S edge. By Lemma 3.2 and 3.3, there are at most $(t+1)^{4km-4l}$ different ways to appoint the regular edges of S to their positions. Here we should note that whether an S-edge is singular or regular is determined by all the edges before it.

From the above arguments and (4.8), we get

$$|E_n| \leq \sum_{\ell=1}^{mk} {2km \choose 2\ell} (k+1)^{2km-2\ell+2m} n^{\ell+1} \sum_{t=1}^{2km-2\ell} {(2km)^2 \choose t} t^{2km-2\ell}$$

$$\times (t+1)^{4km-4l} m^t (\delta \sqrt{n})^{2km-2l-t}$$

$$\leq n^{km+1} \sum_{\ell=1}^{mk} {2km \choose 2\ell} (k+1)^{2km-2\ell+2m} \sum_{t=1}^{2km-2\ell} (2km)^{3t} (t+1)^{6km-6\ell} \delta^{2km-2\ell} (\delta\sqrt{n})^{-t}.$$
 here
$$\sum_{t=1}^{n} A_t = 1, \text{ conveniented for saving notations.}$$

By the elementary inequality

$$a^{t}(t+1)^{b} \le (-\frac{b}{\log a})^{b}$$
 for $(0 < a < 1, b > 0)$

we get

If we select $m = m(n) = A(n) \log n$ such that

1.
$$A(n) \rightarrow \infty$$

2. A(n)
$$\delta^{1/6} \to 0$$

then

$$\frac{6km\delta^{1/6}}{\log \frac{\delta\sqrt{n}}{(2km)^3}} \longrightarrow 0, \quad (n \to \infty).$$

Thus we obtain for large n

$$|E_{n}| \leq n^{km+2} \sum_{\ell=1}^{mk} {2k \choose 2\ell} ((k+1)^{2} \delta)^{km-\ell} (k+1)^{2m}$$

$$\leq n^{km+2} (1 + (k+1) \delta^{1/2})^{2km} (k+1)^{2m}$$

Since z > (1+k) and $\delta \rightarrow 0$, we have

$$\sum_{n=1}^{\infty} z^{-2m} n^{-km} | E_n |$$

$$\leq C \sum_{n=1}^{\infty} (n^{2/m} (1 + (k+1) \delta^{1/2})^{2k} (k+1) / z_0)^m$$

$$\leq C \sum_{n=1}^{\infty} n^m < \infty$$

where $0 < \eta < 1$ is a constant. Here the last series converges because $m/\log n \to \infty$. The proof is finished.

5. TWO PROBLEMS OF GEMAN-HWANG

In Geman-Hwang (1982), they suggested the following system of linear equations with unknown $n \times 1$ vector X_n

$$X_{n} = 1_{n} + \frac{1}{\sqrt{n}} W_{n} X_{n}$$
 (5.1)

where W_n is an $n \times n$ matrix whose (i,j)-entry is w_{ij} and $W = \{w_{ij}: i,j = 1,2,...\}$ is an infinite matrix of iid random variables, and l_n is the $n \times 1$ vector $(1,1,...,1)^T$.

If $X_n = (X_{n1}, ..., X_{nn})^T$, then for any integer $m \ge 1$, Geman and Hwang proved that as $n \to \infty$,

$$(X_{n1}, X_{n2}, \dots, X_{nm})^{T} \longrightarrow N(1_{m}, \frac{\sigma^{2}}{1-\sigma^{2}} I_{m})$$
 weakly, (5.2)

under the conditions

1.
$$E w_{11} = 0$$
, $0 < E w_{11}^2 = \sigma^2 < \frac{1}{4}$;

2. $E|W_{11}^n| \le n^{\alpha n}$ for any integer $n \ge 1$; α is a positive constant.

Geman and Hwang pointed out that the computer simulations support (5.2) even in the case of uniform distribution on [-1,1], where $\sigma^2 = \frac{1}{3}.$

We will prove that (5.2) is true even when $\sigma^2 < 1$ and $E|w_{11}^4| < \infty$.

Theorem 5.1. Let X_n be the solution of (5.1) whenever $(I - \frac{1}{\sqrt{n}} W_n)$ is nonsingular, otherwise define $X_n = 0$. Then (5.2) holds when $E W_{11} = 0$, $E W_{11}^2 = \sigma^2 < 1$ and $E |W_{11}^4| < \infty$.

Geman and Hwang (1982) suggested a system of differential equations

$$\dot{X}_{n}(t) = \alpha X_{n}(t) + \frac{1}{\sqrt{n}} W_{n} X_{n}(t), X_{n}(0) = 1_{n}.$$
 (5.3)

They proved that for any integer $m \ge 1$, real T > 0, $X_{n1}(.), ..., X_{nm}(.)$ (the first m components of the vector $X_n(.)$, the solution of (5.3)) tend to m iid Gaussian processes weakly, as $n \to \infty$, on [0,T]. Each of these m processes has mean $\mu(t) = e^{\alpha t}$ and covariance function

$$C(t,s) = e^{\alpha(t+s)} \sum_{k=1}^{\infty} \frac{(ts)^k}{(k!)^2}$$
.

They supposed among others the following moment requirement

$$E|w_{11}|^n \le n^{\beta n}$$
 for all $n \ge 2$, and some $\beta > 0$.

In the same paper, they conjectured that the analogous theorem should hold for the equation

$$\dot{X}_{n}(t) = \alpha X_{n}(t) + \frac{Wn}{\sqrt{n}} X_{n}(t) + 1_{n}, X_{n}(0) = 1_{n}.$$
 (5.4)

We will prove

Theorem 5.2. Suppose E $w_{11} = 0$, E $w_{11}^2 = 1$, and E $w_{11}^4 < \infty$. Let $X_n(t)$ be the solution of

$$\dot{X}_{n}(t) = \alpha X_{n}(t) + \frac{1}{\sqrt{n}} W_{n} X_{n}(t) + \beta I_{n}, \quad X_{n}(0) = I_{n}. \quad (5.5)$$

Then for any integer $m \ge 1$, real T > 0, $X_{n1}(t), \dots, X_{nm}(t)$ tend to m fid Gaussian processes weakly on $\{0,T\}$ as $n \to \infty$. The mean of these processes is

$$\mu(t) = e^{\alpha t} + \beta \int_0^t e^{\alpha s} ds = e^{\alpha t} + \frac{\beta}{\alpha} (e^{\alpha t} - 1), \qquad (5.6)$$

the covariance function is

$$C(t,s) = \sum_{k=1}^{\infty} \frac{1}{(k!)^2} (t^k e^{\alpha t} + \beta \int_0^t u^k e^{\alpha u} du) (s^k e^{\alpha s} + \beta \int_0^s u^k e^{\alpha u} du).$$
(5.7)

Remark. When β = 0, Theorem 5.2 reduces to an extension of Geman-Hwang theorem. When β = 1, Theorem 5.2 includes a proof of Geman-Hwang's conjecture.

6. PROOF OF THEOREM 5.1

By the Truncation Lemma; we can assume that the entries of W_n are bounded by $\sqrt{n}\delta$, here $\delta=\delta_n \to 0$ arbitrarily slow. We suppose δ is defined as in the proof of Theorem 2.1.

Write Y = $X_n - 1_n$, A = W_n / \sqrt{n} . (5.1) is equivalent to

$$(I_n - A)Y = AI_n$$

Multiply both sides by $\sum_{i=0}^{k-1} A^i$, we get

$$z_n = \frac{\det}{(I_n - A^k)Y} = \sum_{i=1}^k A^i 1_n.$$
 (6.1)

We need the following lemma.

Lemma 6.1. Suppose

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1. $\{w_{ij}; i,j = 1,2,...\}$ are iid random variables; and W_n is the matrix $(w_{ij}; 1 \le i, j \le n)$;

2.
$$E w_{11} = 0$$
, $E w_{11}^2 = \sigma^2$, $E w_{11}^4 < \infty$.

Then if $\alpha(i,k,n)$ denotes the i-th component of the vector $(\frac{W_n}{\sqrt{n}})^k 1_n$, for any distinct ordered pairs $(i_1,k_1),\ldots,(i_m,k_m)$, as $n+\infty$,

$$(\alpha(i_1,k_1,n),\ldots,\alpha(i_m,k_m,n))^T \xrightarrow{W} N_m(0,\Lambda_m),$$

where $\Lambda_{m} = \text{diag}(\sigma^{2k_1}, \dots, \sigma^{2k_m})$.

The proof of Lemma 6.1 is almost the same as the proof in the Appendix of Geman-Hwang (1982) and is therefore omitted here.

By Truncation Lemma and Lemma 6.1, it is not difficult to see that

$$(I_m 0) Z_n \xrightarrow{w} N_m (0, \sum_{i=1}^k \sigma^{2i} I_m), \text{ as } n \to \infty.$$
 (6.2)

Here I_m is the $m \times m$ identity matrix. Also, if $(Z_n)_i$ is the i-th component of Z_n , $E(Z_n)_i^2 \longrightarrow \sum\limits_{i=1}^k \sigma^{2i}$ as $n \to \infty$. Here the reader has to note that we have truncated the entries of W_n at $\sqrt{n} \delta$.

In order to prove Theorem 5.1, we notice that

$$\begin{aligned} & X_n = 1_n + Y = 1_n + Z_n + A^k Y. \\ & \text{Then, if } t = (t_1, \dots, t_m)^T, \ i = \sqrt{-1}, \\ & | E \ e^{it'(I_m 0)(X_n - 1_n)} - e^{-1/2 \ t't \sum_{j=1}^{\infty} \sigma^{2j}} | \\ & \leq \left| E \ e^{it'(I_m 0)(X_n - 1_n)} - E \ e^{it'(I_m 0)Z_n} \right| \\ & + \left| E \ e^{it'(I_m 0)Z_{n-}} \exp\{-\frac{1}{2} \ t't \sum_{j=1}^{k} \sigma^{2j}\} \right| + \left| \exp\{-\frac{1}{2} \ t't \sum_{j=1}^{k} \sigma^{2j}\} - \exp\{-\frac{1}{2} t't \sum_{j=1}^{\infty} \sigma^{2j}\} \right| + \left| \exp\{-\frac{1}{2} \ t't \sum_{j=1}^{k} \sigma^{2j}\} - \exp\{-\frac{1}{2} t't \sum_{j=1}^{\infty} \sigma^{2j}\} \right| + \left| \exp\{-\frac{1}{2} \ t't \sum_{j=1}^{k} \sigma^{2j}\} - \exp\{-\frac{1}{2} t't \sum_{j=1}^{\infty} \sigma^{2j}\} \right| + \left| \exp\{-\frac{1}{2} \ t't \sum_{j=1}^{k} \sigma^{2j}\} - \exp\{-\frac{1}{2} t't \sum_{j=1}^{\infty} \sigma^{2j}\} \right| + \left| \exp\{-\frac{1}{2} \ t't \sum_{j=1}^{k} \sigma^{2j}\} - \exp\{-\frac{1}{2} t't \sum_{j=1}^{\infty} \sigma^{2j}\} \right| + \left| \exp\{-\frac{1}{2} \ t't \sum_{j=1}^{k} \sigma^{2j}\} - \exp\{-\frac{1}{2} t't \sum_{j=1}^{\infty} \sigma^{2j}\} \right| + \left| \exp\{-\frac{1}{2} \ t't \sum_{j=1}^{\infty} \sigma^{2j}\} - \exp\{-\frac{1}{2} t't \sum_{j$$

As $n \to \infty$, $a_2 \to 0$, by (6.2).

 $= a_1 + a_2 + a_3,$

Now we estimate a_1 . We have for any $\epsilon > 0$

$$a_1 \le E[e^{it'(I_m0)A^kY} - 1] \le 2P([|(I_m0)A^kY|] \ge \epsilon) + \phi(\epsilon)$$

Here
$$\phi(\varepsilon) = \sup_{||\mathbf{x}|| \le \varepsilon} |e^{\mathbf{t}'\mathbf{x}} - 1| \to 0 \text{ as } \varepsilon \to 0.$$

We consider only those k, for which $(1+k)^{1/k} \sigma < 1$.

Let
$$\Delta = \Delta_{n,k} = \{\omega \in \Omega: ||A^k|| < \eta^k\}, \text{ where } (1+k)^{1/k} \sigma < \eta < 1,$$

n is fixed. Evidently $P(\Delta) \rightarrow 1$ as $n \rightarrow \infty$ by Theorem 2.1. Thus

$$P(||(I_{m}0)A^{k}Y|| \geq \varepsilon) \leq P(||(I_{m}0)A^{k}Y|| \geq \varepsilon, ||A^{k}|| < \eta^{k}) + P(||A^{k}|| \geq \eta^{k})$$

$$\leq \frac{1}{\varepsilon^{2}} E||(I_{m}0)A^{k}Y||^{2} 1_{\Delta} + P(||A^{k}|| \geq \eta^{k})$$

$$\leq \frac{m}{\varepsilon^{2}} E||A^{k}Y||^{2} 1_{\Delta} + 1 - P(\Delta), \qquad (6.3)$$

since the components of $A^{k}Y$ 1_{Δ} have the same distribution.

We have

$$A^{k}Y = A^{k}(I - A^{k})Y + A^{k}A^{k}Y = A^{k}Z_{n} + A^{k}(A^{k}Y),$$

so

$$||A^{k}Y|| \le ||A^{k}|| ||Z_{n}|| + ||A^{k}|| ||A^{k}Y||,$$

and

$$||A^{k}Y|| 1_{\Delta} \le \frac{||A^{k}||}{1 - ||A^{k}||} ||Z_{n}|| 1_{\Delta} \le \frac{n^{k}}{1 - n^{k}} ||Z_{n}|| 1_{\Delta}$$
 (6.4)

By (6.3) and (6.4),

$$P(\left|\left|\left(I_{\underline{m}}0\right)A^{\underline{k}}Y\right|\right| \geq \epsilon) \leq \frac{\underline{m}}{\epsilon^{2}\underline{n}} \left(\frac{\underline{n}^{\underline{k}}}{1-\underline{n}^{\underline{k}}}\right)^{2} E\left|\left|Z_{\underline{n}}\right|\right|^{2} + 1 - P(\Delta).$$

Let $n \rightarrow \infty$, we get

$$\frac{\overline{\lim}}{n+\infty} \quad P(||(\underline{I}_{\underline{m}}0)\mathbf{A}^{\underline{k}}\mathbf{Y}|| \geq \varepsilon) \leq \frac{1}{\varepsilon^2} \left(\frac{\eta^{\underline{k}}}{1-\eta^{\underline{k}}}\right)^2 \quad \sum_{\underline{j=1}}^{\underline{k}} \sigma^{2\underline{j}}.$$

So

$$\frac{\overline{\lim_{n\to\infty}} |E| e^{it'(I_m 0)(X_n - I_n)} - \exp\{-\frac{1}{2}t't |\int_{j=1}^{\infty} \sigma^{2j}\}| \leq \frac{\overline{\lim}}{n\to\infty} a_1 + a_3$$

$$\leq \frac{1}{\epsilon^{2}} \left(\frac{n^{k}}{1-n^{k}} \right)^{2} \sum_{j=1}^{k} \sigma^{2j} + \phi(\epsilon) + \left| \exp\{-\frac{1}{2} t' t \int_{j=1}^{k} \sigma^{2j} \} - \exp\{-\frac{1}{2} t' t \int_{j=1}^{\infty} \sigma^{2j} \} \right|$$

Letting $k \to \infty$, and then $\epsilon \to 0$, we see that the left hand side tends to zero.

PROOF OF THEOREM 5.2

It is easy to verify that

$$X_n(t) = \sum_{k=0}^{\infty} \frac{1}{k!} \left(\frac{W_n}{\sqrt{n}} \right)^k 1_n \left(t^k e^{\alpha t} + \beta \int_0^t s^k e^{\alpha s} ds \right)$$
 (7.1)

is the solution to (5.5).

Theorem 5.2 is a consequence of the following lemma.

Lemma 7.1. Let $\{w_{ij}: i,j=1,2,...\}$ be a family of iid random variables with E $w_{11}=0$, E $w_{11}^2=1$ and E $w_{11}^4<\infty$, and $w_n=(w_{ij},1)$, $1 \le i \le n$, $1 \le i \le n$.

Let $\{g_k(.), k = 0,1,...\}$ be a sequence of continuous functions satisfying

$$\sum_{k=0}^{\infty} \frac{r^{k}}{k!} \sup_{0 \le t \le T} |g_{k}(t)| < \infty, \qquad (7.2)$$

here r > 2, T > 0 are positive constants.

Then for any integer $m \ge 1$, as $n \to \infty$ the stochastic process $(I_m 0) \sum_{k=0}^{\infty} \frac{1}{k!} \left(\frac{Wn}{\sqrt{n}} \right)^k 1_n g_k(t), \ t \in [0,T], \ \text{tends to an } m\text{-dimensional}$ Gaussian process with iid components, each with mean $g_0(t)$ and covariance function $c(t,s) = \sum_{k=1}^{\infty} \left(\frac{1}{k!} \right)^2 g_k(t) g_k(s)$.

Proof. Let

$$Z_n(t) = (Z_{n1}(t), \dots, Z_{nn}(t))^T = \sum_{k=1}^{\infty} \frac{1}{k!} (\frac{w_n}{\sqrt{n}})^k I_n g_k(t).$$

We prove that the sequence $\{(Z_{n1}(.),...,Z_{nm}(.)), n=1,2,...\}$ of stochastic processes is tight in $C^m[0,t]$. It is easy to see that we need only to show that $\{Z_{ni}(.), n=1,2,...\}$ is tight in $C[0,T], 1 \le i \le m$. Let $\Delta_n = \{\omega \in \Omega \colon \left|\left|\frac{Wn}{\sqrt{n}}\right| \mid (\omega) \le r\}$. By Theorem 2.1, $P(\Delta_n) \to 1$.

Let

$$\rho_{\mathbf{k}}(\delta) = \sup_{\substack{|\mathbf{t}-\mathbf{s}| < \delta \\ \mathbf{t}, \mathbf{s} \in [0,T]}} |\mathbf{g}_{\mathbf{k}}(\mathbf{t}) - \mathbf{g}_{\mathbf{k}}(\mathbf{s})|,$$

$$\alpha(\mathbf{i}, \mathbf{k}, \mathbf{n}) = \{(\frac{\mathbf{W}_{\mathbf{n}}}{\sqrt{\mathbf{n}}})^{\mathbf{k}} \mathbf{1}_{\mathbf{n}}\}_{\mathbf{i}} = \text{the i-th component of } (\frac{\mathbf{W}_{\mathbf{n}}}{\sqrt{\mathbf{n}}})^{\mathbf{k}} \mathbf{1}_{\mathbf{n}}.$$

We have

$$\sup_{\substack{|t-s|<\delta\\t,s\in[0,T]}} |Z_{ni}(t)-Z_{ni}(s)| \leq \sum_{k=1}^{\infty} |\alpha(i,k,n)| \frac{\rho(\delta)}{k!},$$

hence

$$\lim_{\delta \to 0} \frac{\overline{\lim}}{n \to \infty} \quad P(\sup_{\substack{t-s \\ t,s \in T}} |Z_{ni}(t) - Z_{ni}(s)| > \varepsilon) \\
\leq \lim_{\delta \to 0} \frac{\overline{\lim}}{n \to \infty} \quad P(\sum_{k=1}^{\infty} |\alpha(i,k,n)| \frac{\rho_{k}(\delta)}{k!} > \varepsilon) \\
\leq \lim_{\delta \to 0} \frac{\overline{\lim}}{n \to \infty} \left[\frac{1}{\varepsilon} E \sum_{k=1}^{\infty} \frac{1}{n} |\alpha(i,k,n)| \frac{\rho_{k}(\delta)}{k!} + (1 - P(\Delta_{n})) \right] \\
= \lim_{\delta \to 0} \frac{\overline{\lim}}{n \to \infty} \frac{1}{\varepsilon} \sum_{k=1}^{\infty} \frac{\rho_{k}(\delta)}{k!} \quad E \mid_{\Delta_{n}} |\alpha(i,k,n)|.$$

It is easy to see that $\alpha(i,k,n)$ $1_{\Delta_n},\ldots,\alpha(n,k,n)$ $1_{\Delta_n},$ $i=1,2,\ldots,n$, have an identical distribution. Therefore

$$\begin{split} & \mathbb{E} \, \mathbf{1}_{\Delta_{\mathbf{n}}} \, \left| \alpha(\mathbf{i}, \mathbf{k}, \mathbf{n}) \right| \leq \mathbb{E}^{1/2} \, \mathbf{1}_{\Delta_{\mathbf{n}}} \, \left| \alpha(\mathbf{i}, \mathbf{k}, \mathbf{n}) \right|^{2} \\ & \leq \left[\frac{1}{n} \, \mathbb{E} \, \mathbf{1}_{\Delta_{\mathbf{n}}} \, \left| \left| \left(\frac{\mathbf{W}_{\mathbf{n}}}{\sqrt{\mathbf{n}}} \right)^{k} \, \mathbf{1}_{\mathbf{n}} \right| \right|^{2} \right]^{1/2} \\ & \leq \left[\mathbb{E} \, \mathbf{1}_{\Delta_{\mathbf{n}}} \, \left| \left| \frac{\mathbf{W}_{\mathbf{n}}}{\sqrt{\mathbf{n}}} \right| \right|^{2k} \right]^{1/2} \leq r^{k}. \end{split}$$

So,

$$\frac{\lim \overline{\lim} \ \overline{\lim}}{\delta + 0 \ n \to \infty} P(\sup_{\substack{|\mathbf{t} - \mathbf{s}| < \delta \\ \mathbf{t}, \mathbf{s} \in T}} |Z_{ni}(\mathbf{t}) - Z_{ni}(\mathbf{s})| > \varepsilon)$$

$$\leq \lim_{\delta \to 0} \frac{1}{\varepsilon} \sum_{k=1}^{\infty} \frac{\rho_{k}(\delta)}{k!} r^{k} = 0.$$

Thus, the tightness of the family $\{Z_{ni}(.); n = 1,2,...\}$ of stochastic processes is established.

Finally we show that for any positive integer ℓ and

$$t_1, \ldots, t_{\ell} \in [0,T], \text{ as } n \rightarrow \infty$$

$$E \exp\{i \sum_{\nu=1}^{m} \sum_{j=1}^{\ell} \lambda_{\nu j} z_{n\nu}(t_{j})\} \longrightarrow \exp\{-\frac{1}{2} \sum_{\nu=1}^{m} \sum_{j=1}^{\ell} \sum_{q=1}^{\ell} \lambda_{\nu j} \lambda_{\nu q} C(t_{j}, t_{q})\}$$

here $i = \sqrt{-1}$ and $\{\lambda_{v,j}\}$ are real numbers.

Let

$$e_{n\nu}^{p}(t) = \sum_{k=p+1}^{\infty} \left(\left(\frac{w_{n}}{\sqrt{n}} \right)^{k} 1_{n} \right)_{\nu} \frac{g_{k}(t)}{k!}$$

$$= \sum_{k=p+1}^{\infty} \alpha(\nu, k, n) \frac{g_{k}(t)}{k!}, \quad \nu = 1, \dots, n.$$

Let $g_k = \sup_{t \in [0,T]} g_k(t)$. Then for any $\epsilon > 0$,

$$\lim_{p\to\infty} \frac{\lim}{n\to\infty} P(|e_{\vee n}^{p}(t_{j})| \ge \varepsilon) \le \lim_{p\to\infty} \frac{\lim}{n\to\infty} \frac{1}{\varepsilon} \sum_{k=p+1}^{\infty} \frac{g_{k}}{k!} E 1_{\Delta_{n}} |\alpha(\vee,k,n)|$$

$$\le \frac{1}{\varepsilon} \lim_{p\to\infty} \sum_{k=p+1}^{\infty} \frac{r^{k}}{k!} g_{k} = 0.$$
(7.3)

On the other hand, by (7.2)

$$\frac{\overline{\lim}}{|\sum_{p\to\infty}^{\infty}\frac{1}{|k=p+1|}\frac{1}{(k!)^2}}g_k(t_j)g_k(t_q)| \leq \frac{\overline{\lim}}{|\sum_{p\to\infty}^{\infty}\frac{g_k}{|k=p+1|}\frac{g_k}{(k!)}}^2 = 0. \quad (7.4)$$

We have

$$\begin{aligned} & \left| \mathbf{E} \, \exp \left\{ i \, \sum_{\nu=1}^{m} \, \sum_{j=1}^{k} \lambda_{\nu,j} \, \sum_{k=1}^{\infty} \alpha(\nu, \, k, \, n) \, \frac{\mathbf{g}_{k}(\mathbf{t}_{j})}{k!} \right\} \\ & - \mathbf{E} \, \exp \left\{ -\frac{1}{2} \, \sum_{\nu=1}^{m} \, \sum_{j=1}^{k} \sum_{q=1}^{k} \lambda_{\nu,j} \, \lambda_{\nu,q} \mathbf{c}(\mathbf{t}_{j}, \, \mathbf{t}_{q}) \right\} \right| \\ & \leq \left| \mathbf{E} \, \exp \left\{ i \, \sum_{\nu=1}^{m} \, \sum_{j=1}^{k} \lambda_{\nu,j} \, \sum_{k=1}^{\infty} \alpha(\nu, \, k, \, n) \, \frac{\mathbf{g}_{k}(\mathbf{t}_{j})}{k!} \right\} \\ & - \mathbf{E} \, \exp \left\{ i \, \sum_{\nu=1}^{m} \, \sum_{j=1}^{k} \lambda_{\nu,j} \, \sum_{k=1}^{p} \alpha(\nu, \, k, \, n) \, \frac{\mathbf{g}_{k}(\mathbf{t}_{j})}{k!} \right\} \right| \end{aligned}$$

$$+ \left| E \exp \left\{ i \sum_{v=1}^{m} \sum_{j=1}^{k} \lambda_{vj} \sum_{k=1}^{p} \alpha(v, k, n) \frac{g_{k}(t_{j})}{k!} \right\}$$

$$- \exp \left\{ -\frac{1}{2} \sum_{v=1}^{m} \sum_{j=1}^{p} \left(\sum_{j=1}^{k} \lambda_{vj} \frac{g_{k}(t_{j})}{k!} \right)^{2} \right\} \right|$$

$$+ \left| \left\{ \exp \left\{ -\frac{1}{2} \sum_{v=1}^{m} \sum_{k=1}^{p} \left(\sum_{j=1}^{k} \lambda_{vj} \frac{g_{k}(t_{j})}{k!} \right)^{2} \right\} \right|$$

$$- \exp \left\{ -\frac{1}{2} \sum_{v=1}^{m} \sum_{j=1}^{k} \sum_{q=1}^{k} \lambda_{vj} \lambda_{vq} c(t_{j}, t_{q}) \right\} \right|$$

$$= a_{1} + a_{2} + a_{3}.$$

By (7.3) $\overline{\lim} \overline{\lim} a_1 = 0$. By Lemma 6.1, $\lim_{n\to\infty} a_2 = 0$. And

$$a_{3} = \left| \exp \left(-\frac{1}{2} \sum_{v=1}^{m} \sum_{j=1}^{\ell} \sum_{q=1}^{\ell} \lambda_{vj} \lambda_{vq} \sum_{k=1}^{p} \left(\frac{1}{k!}\right)^{2} g(t_{j}) g(t_{q}) \right\}$$

$$- \exp \left\{-\frac{1}{2} \sum_{v=1}^{m} \sum_{j=1}^{\ell} \sum_{q=1}^{\ell} \lambda_{vj} \lambda_{vq} \sum_{k=1}^{\infty} \left(\frac{1}{k!}\right)^{2} g(t_{j}) g(t_{q}) \right\} \right|$$

$$\leq \left| 1 - \exp \left\{-\frac{1}{2} \sum_{v=1}^{m} \sum_{j=1}^{\ell} \sum_{q=1}^{\ell} \lambda_{vj} \lambda_{vq} \sum_{k=p+1}^{\infty} \left(\frac{1}{k!}\right)^{2} g(t_{j}) g(t_{q}) \right\} \right|$$

$$\times \left| \exp \left\{\frac{1}{2} \sum_{v=1}^{m} \sum_{j=1}^{\ell} \sum_{q=1}^{\ell} \lambda_{vj} ||\lambda_{vq}| \left(\sum_{k=1}^{\infty} \frac{g_{k}}{k!}\right)^{2} \right\} \longrightarrow 0, \text{ as } p + \infty,$$

by (7.4). We finish the proof.

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where W:n x n is a square random matrix with i.i.d. entries and σ^2 is the variance of the entries of W. In proving the result, the authors assumed the existance of fourth moment of the entries of W.			
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